

Data Analysis on DmartReady Online Store

Introduction

The DmartReady Online Store Analysis is a comprehensive data-driven project aimed at understanding the operational, financial, and customer behavioral patterns of DmartReady, an online retail platform. The analysis leverages SQL and Python to extract, transform, and analyze large datasets, providing valuable insights into sales trends, customer demographics, product performance, and operational efficiency. By examining key metrics such as total sales, order volumes, customer engagement, and marketing performance, the project identifies factors influencing the business's growth and profitability.

Objectives

The primary objectives of this analysis include: **Understanding Sales Performance:** Analyze total and average sales, sales by category, and revenue contribution by gender and marketing campaigns. **Customer Behavior Analysis:** Study customer demographics, average time spent on the website, and buying patterns by state, city, and age group. **Product Analysis:** Identify top-performing products, products with the highest discount rates, and products with the highest customer engagement. **Operational Efficiency:** Evaluate delivery times, ship modes, and cancellation rates across states and cities to improve logistics. **Marketing Insights:** Assess the effectiveness of different marketing platforms in driving sales and revenue. **Profitability Assessment:** Determine profit margins across various product categories.

Data Analysis Workflow: Importing necessary libraries

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import mysql.connector
import plotly.graph_objects as go
import plotly.express as px
```

Data Collection: Connect with SQL database, Extract data using SQL queries.

```
In [204... db = mysql.connector.connect(
    host='127.0.0.1',
    user='root',
    password='1234',
    database='dmart'
```

```
)
cursor = db.cursor() # Creates a cursor object from the established connection
```

In [266...

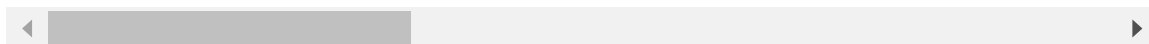
```
query = """ SELECT * from dmart.dmartready"""
cursor.execute(query) # Executes a SELECT query
data = cursor.fetchall() # Fetches all rows from the executed query
df = pd.DataFrame(data, columns = ["CustomerID", "ProductID", "OrderID", "Age",
                                   "DiscountPrice", "Category", "State", "City",
                                   "Rating", "Marketing/Advertisement", "ShipMod
                                   "PaymentMethod", "PinCode", "TotalOrderValue"
                                   "No.of_Clicks", "Year", "Month", "ShippingCha

df.head()
# Converts the fetched SQL data into a pandas DataFrame for further analysis.
```

Out[266...

| | CustomerID | ProductID | OrderID | Age | Gender | ProductName | MRP | DiscountP |
|---|------------|-----------|-----------|-----|--------|--|---------|-----------|
| 0 | YSB75 | BW653 | 479577309 | 28 | Female | Ariel Matic Top Load Liquid Detergent | 977.44 | 90% |
| 1 | FUS93 | XV061 | 634865221 | 47 | Male | VSR Channa Dal | 834.09 | 45% |
| 2 | AJP28 | GF695 | 113166210 | 63 | Female | Tenali Double Horse Chana Dal | 1095.20 | 100% |
| 3 | URC55 | VM478 | 740539230 | 41 | Male | Tata Tea | 748.16 | 38% |
| 4 | ZOP23 | XD230 | 156544145 | 22 | Female | VSR Channa Dal | 1249.04 | 91% |

5 rows × 29 columns



Understanding the data & cleaning:

In [267...

```
# Used to generate descriptive statistics for the numerical columns in a DataFrame
df.describe().round()
```

Out[267...

| | OrderID | Age | MRP | DiscountPrice | BillNumber | TimeSpentonWebsite |
|-------|-------------|---------|---------|---------------|--------------|--------------------|
| count | 25000.0 | 25000.0 | 25000.0 | 25000.0 | 2.500000e+04 | 25000.0 |
| mean | 498712983.0 | 44.0 | 1014.0 | 736.0 | 5.016699e+11 | 10.0 |
| std | 288189483.0 | 15.0 | 571.0 | 441.0 | 2.887199e+11 | 11.0 |
| min | 19128.0 | 18.0 | 20.0 | 11.0 | 2.349470e+07 | 1.0 |
| 25% | 250374576.0 | 31.0 | 524.0 | 368.0 | 2.540238e+11 | 3.0 |
| 50% | 499731127.0 | 44.0 | 1014.0 | 713.0 | 5.019725e+11 | 5.0 |
| 75% | 747102105.0 | 57.0 | 1507.0 | 1061.0 | 7.519322e+11 | 13.0 |
| max | 999913093.0 | 70.0 | 2000.0 | 1876.0 | 9.999160e+11 | 60.0 |



```
In [270... # Checking rows & columns present in data frame
df.shape
```

```
Out[270... (25000, 29)
```

```
In [272... df.columns
```

```
Out[272... Index(['CustomerID', 'ProductID', 'OrderID', 'Age', 'Gender', 'ProductName',
      'MRP', 'DiscountPrice', 'Category', 'State', 'City', 'Subscription',
      'BillNumber', 'TimeSpentonWebsite', 'Rating', 'Marketing/Advertisement',
      'ShipMode', 'OrderStatus', 'OrderDate', 'DeliveryDate',
      'CancellationDate', 'PaymentMethod', 'PinCode', 'TotalOrderValue',
      'PaymentStatus', 'No.of_Clicks', 'Year', 'Month', 'ShippingCharges'],
      dtype='object')
```

```
In [274... # Checking overview of data
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25000 entries, 0 to 24999
Data columns (total 29 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   CustomerID                           25000 non-null  object
1   ProductID                             25000 non-null  object
2   OrderID                               25000 non-null  int64
3   Age                                    25000 non-null  int64
4   Gender                                25000 non-null  object
5   ProductName                           25000 non-null  object
6   MRP                                    25000 non-null  float64
7   DiscountPrice                         25000 non-null  float64
8   Category                              25000 non-null  object
9   State                                 25000 non-null  object
10  City                                   25000 non-null  object
11  Subscription                           25000 non-null  object
12  BillNumber                             25000 non-null  float64
13  TimeSpentonWebsite                     25000 non-null  float64
14  Rating                                 25000 non-null  float64
15  Marketing/Advertisement                 25000 non-null  object
16  ShipMode                               25000 non-null  object
17  OrderStatus                           25000 non-null  object
18  OrderDate                             25000 non-null  object
19  DeliveryDate                           25000 non-null  object
20  CancellationDate                       1784 non-null   object
21  PaymentMethod                          25000 non-null  object
22  PinCode                                25000 non-null  int64
23  TotalOrderValue                        25000 non-null  object
24  PaymentStatus                          25000 non-null  object
25  No.of_Clicks                           25000 non-null  int64
26  Year                                    25000 non-null  int64
27  Month                                  25000 non-null  object
28  ShippingCharges                        25000 non-null  object
dtypes: float64(5), int64(5), object(19)
memory usage: 5.5+ MB
```

```
In [276... # Using Conditional statement checking for duplicate value in dataframe
if df.duplicated().sum() > 0:
    print('Duplicates are present in dataframe')
```

```
else:
    print('Duplicate does not exist')
```

Duplicate does not exist

```
In [278... # Another method of check duplicates
df.duplicated().sum()
```

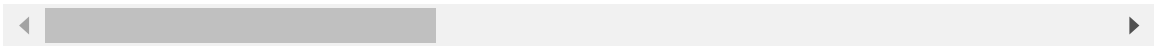
Out[278... 0

```
In [406... df.isna()
```

Out[406...

| | CustomerID | ProductID | OrderID | Age | Gender | ProductName | MRP | DiscountF |
|-------|------------|-----------|---------|-------|--------|-------------|-------|-----------|
| 0 | False | False | False | False | False | False | False | F |
| 1 | False | False | False | False | False | False | False | F |
| 2 | False | False | False | False | False | False | False | F |
| 3 | False | False | False | False | False | False | False | F |
| 4 | False | False | False | False | False | False | False | F |
| ... | ... | ... | ... | ... | ... | ... | ... | |
| 24995 | False | False | False | False | False | False | False | F |
| 24996 | False | False | False | False | False | False | False | F |
| 24997 | False | False | False | False | False | False | False | F |
| 24998 | False | False | False | False | False | False | False | F |
| 24999 | False | False | False | False | False | False | False | F |

25000 rows × 31 columns



```
In [280... # Check for null values
df.isnull().sum()
```

```
Out[280... CustomerID      0
ProductID      0
OrderID        0
Age            0
Gender         0
ProductName    0
MRP            0
DiscountPrice  0
Category       0
State          0
City           0
Subscription   0
BillNumber     0
TimeSpentonWebsite  0
Rating         0
Marketing/Advertisement  0
ShipMode       0
OrderStatus    0
OrderDate      0
DeliveryDate    0
CancellationDate 23216
PaymentMethod   0
PinCode         0
TotalOrderValue 0
PaymentStatus   0
No.of_Clicks    0
Year            0
Month           0
ShippingCharges 0
dtype: int64
```

```
In [20]: df.rename(columns={
        'TotalOrderValue': 'Sales'}, inplace=True)
```

```
In [26]: df['Sales'] = df['Sales'].str.replace('abc|xyz', '', regex=True)
```

```
In [30]: df['Sales'] = df['Sales'].str.strip() # Removes leading and trailing spaces
```

```
In [32]: df['Sales'] = df['Sales'].str.replace(r'[!#@]', '', regex=True) # Remove specific
```

```
In [46]: # Replace "Cancelled" and "Returned" in the 'Order_Status' column with 0
df['Sales'] = df['Sales'].where(~df['Sales'].isin(['Cancelled', 'Returned']), 0)
```

```
In [52]: # Change data type from string to integer
df['Sales'] = df['Sales'].astype(float)
```

Exploratory Data Analysis & Visualization

```
In [200... # Total Sales ?
query = """ select round(sum(total_order_value)) from dmartready"""
cursor.execute(query)
data1a = cursor.fetchall()
data1a
```

```
Out[200... [(16954109.0,)]
```

In [410... `df['TotalOrderValue'].sum().round()`

Out[410... 16954109.0

In [184... *# Find sales contribution by gender?*
Customer Segmentation
`df['AgeRange'] = pd.cut(df['Age'], bins=[18, 25, 35, 45], labels=["18-25", "26-35", "36-45"])`
`customer_segmentation = df.groupby(['Gender', 'AgeRange'])['Sales'].sum()`
`customer_segmentation`

Out[184...

| Gender | AgeRange | Sales |
|--------|----------|------------|
| Female | 18-25 | 958540.39 |
| | 26-35 | 1362935.26 |
| | 36-45 | 1412049.54 |
| Male | 18-25 | 1272582.93 |
| | 26-35 | 1867735.05 |
| | 36-45 | 1870226.81 |

 Name: Sales, dtype: float64

In [194... `data = {`
 `'Gender': ['Female', 'Female', 'Female', 'Male', 'Male', 'Male'],`
 `'AgeRange': ['18-25', '26-35', '36-45', '18-25', '26-35', '36-45'],`
 `'Sales': [958540.39, 1362935.26, 1412049.54, 1272582.93, 1867735.05, 1870226.81]`
`}`
`df = pd.DataFrame(data)`
`fig = px.sunburst(`
 `df,`
 `path=['Gender', 'AgeRange'],`
 `values='Sales',`
 `color='Gender',`
 `template="plotly_dark",`
 `title="Customer Segmentation by Gender and Age Range"`
`)`
`fig.show()`

```
In [62]: UniqueOrders = df['OrderID'].nunique()
print(UniqueOrders)
```

25000

```
In [54]: # what is the average sales ?
AverageSalesPerOrder = df.groupby('OrderID')['Sales'].sum().mean()
AverageSalesPerOrder = round(AverageSalesPerOrder, 2)

# Output the result with formatting
print(f"Average Sales per Order: ₹{AverageSalesPerOrder:,.2f}")
```

Average Sales per Order: ₹678.16

```
In [56]: # Get the total number of products sold in each category.
query = """ SELECT Category, COUNT(Product_ID) AS TotalProductsSold
FROM dmartready GROUP BY Category"""
cursor.execute(query)
data1 = cursor.fetchall()
df1 = pd.DataFrame(data1, columns = ["Category", "TotalProductSold"])
df1.head()
```

Out[56]:

| | Category | TotalProductSold |
|---|----------|------------------|
| 0 | Imported | 5034 |
| 1 | Branded | 7564 |
| 2 | Local | 12402 |

```
In [550... Category = ['Imported', 'Branded', 'Local']
TotalProductSold = [5034, 7564, 12402]

fig = go.Figure(data=[go.Pie(labels=Category, values=TotalProductSold, hole=0.5)
fig.update_traces(textinfo='percent+label', pull=[0, 0, 0.1]) # Set the layout
# Add title and labels
fig.update_layout(
    title="Pie Chart of Product Categories",
    title_x=0.5,
    template="plotly_dark",
    showlegend=True
)
fig.show()
```

```
In [126... # Get the number of orders placed in each month for a specific year
query = """SELECT Month, COUNT(Order_ID) AS OrdersInMonth
FROM dmartready
WHERE Year = 2023
GROUP BY Month """
```



```

cursor.execute(query)
data1 = cursor.fetchall()
df1 = pd.DataFrame(data1, columns = ["Month", "OrdersInMonth2023"])
# correct month order
month_order = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'September', 'October', 'November', 'December']
# Convert the 'Month' column to a categorical type with the specified order
df1['Month'] = pd.Categorical(df1['Month'], categories=month_order, ordered=True)
df_sorted1 = df1.sort_values(by=['Month'])

```

In [124...

```

query = """SELECT Month, COUNT(Order_ID) AS OrdersInMonth
FROM dmartready
WHERE Year = 2022
GROUP BY Month """
cursor.execute(query)
data2 = cursor.fetchall()
df2 = pd.DataFrame(data2, columns = ["Month", "OrdersInMonth2022"])
# correct month order
month_order = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'September', 'October', 'November', 'December']
# Convert the 'Month' column to a categorical type with the specified order
df2['Month'] = pd.Categorical(df2['Month'], categories=month_order, ordered=True)
df_sorted2 = df2.sort_values(by=['Month'])

```

In [122...

```

query = """SELECT Month, COUNT(Order_ID) AS OrdersInMonth
FROM dmartready
WHERE Year = 2021
GROUP BY Month """
cursor.execute(query)
data3 = cursor.fetchall()
df3 = pd.DataFrame(data3, columns = ["Month", "OrdersInMonth2021"])
# correct month order
month_order = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'September', 'October', 'November', 'December']
# Convert the 'Month' column to a categorical type with the specified order
df3['Month'] = pd.Categorical(df3['Month'], categories=month_order, ordered=True)
df_sorted3 = df3.sort_values(by=['Month'])

```

In [116...

```

# Merge the DataFrames on the 'Month' column
MergedData = pd.merge(df1, df2, on="Month", how="outer")
MergedData = pd.merge(MergedData, df3, on="Month", how="outer")
MergedData

```

Out[116...

| | Month | OrdersInMonth2023 | OrdersInMonth2022 | OrdersInMonth2021 |
|----|-----------|-------------------|-------------------|-------------------|
| 0 | January | 678 | 644 | 681 |
| 1 | February | 680 | 651 | 627 |
| 2 | March | 711 | 706 | 715 |
| 3 | April | 665 | 686 | 715 |
| 4 | May | 722 | 709 | 689 |
| 5 | June | 660 | 697 | 655 |
| 6 | July | 687 | 722 | 711 |
| 7 | August | 706 | 765 | 711 |
| 8 | September | 684 | 620 | 658 |
| 9 | October | 723 | 704 | 714 |
| 10 | November | 711 | 688 | 750 |
| 11 | December | 738 | 733 | 684 |

In [548...

```
# Create bar chart
fig = px.bar(MergedData, x="Month", y=["OrdersInMonth2021", "OrdersInMonth2022",
    title="Orders Comparison Across Years (2021, 2022, 2023)",
    labels={"value": "Orders", "variable": "Year"},
    template="plotly_dark")
fig.show()
```

```
In [106... query = """select year, count(order_id) as OrdersInYear
from dmartready
group by year """
cursor.execute(query)
data4 = cursor.fetchall()
df4 = pd.DataFrame(data4, columns = ["Year", "OrdersInYear"])
df4_sort = df4.sort_values(by=['Year'])
df4_sort
```

```
Out[106...   Year  OrdersInYear
1  2021           8310
0  2022           8325
2  2023           8365
```

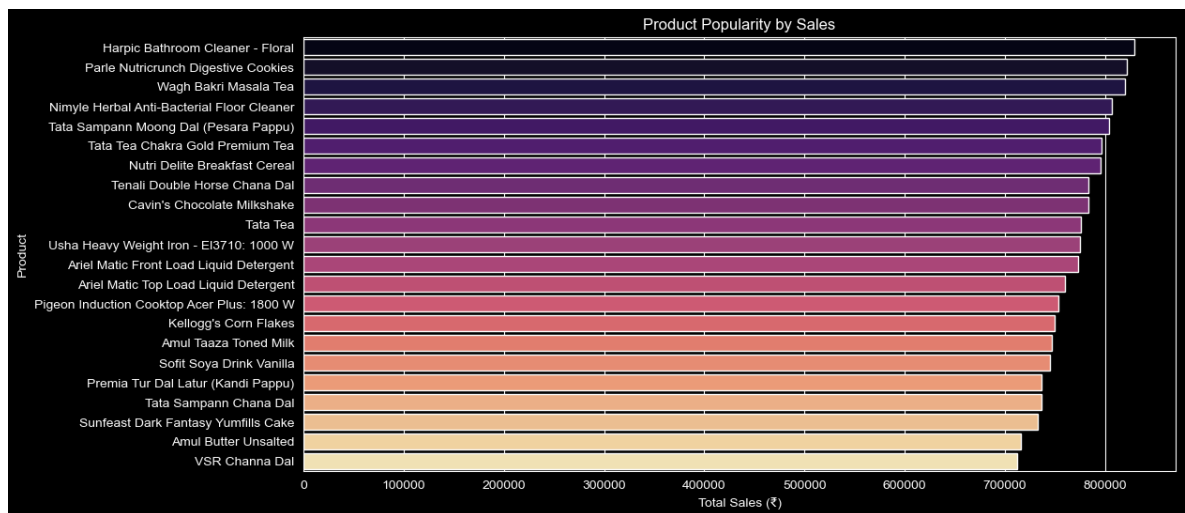
```
In [414... # Calculate the total revenue generated from each product
query = """SELECT Product_Name, round(SUM(Total_Order_Value)) AS TotalRevenue
FROM dmartready
GROUP BY Product_Name order by TotalRevenue desc"""
cursor.execute(query)
data5 = cursor.fetchall()
df5 = pd.DataFrame(data5, columns = ["ProductName", "TotalSales"])
df5.head()
```

Out[414...

| | ProductName | TotalSales |
|---|--|------------|
| 0 | Harpic Bathroom Cleaner - Floral | 828936.0 |
| 1 | Parle Nutricrunch Digestive Cookies | 821713.0 |
| 2 | Wagh Bakri Masala Tea | 820168.0 |
| 3 | Nimyle Herbal Anti-Bacterial Floor Cleaner | 807119.0 |
| 4 | Tata Sampann Moong Dal (Pesara Pappu) | 804085.0 |

In [386...

```
# Product Popularity
plt.style.use('dark_background')
product_popularity = df5.groupby('ProductName')['TotalSales'].sum().sort_values()
plt.figure(figsize=(12, 6))
sns.barplot(x=product_popularity.values, y=product_popularity.index, palette='magma')
plt.title('Product Popularity by Sales')
plt.xlabel('Total Sales (₹)')
plt.ylabel('Product')
plt.show()
```



In [418...

```
# Determine the average rating for products in each product_name
query = """SELECT product_name, round(AVG(Rating),2) AS AverageRating
FROM dmartready
GROUP BY product_name order by AverageRating desc"""
cursor.execute(query)
data5a = cursor.fetchall()
df5a = pd.DataFrame(data5a, columns = ["ProductName", "AverageRating"])
df5a.head()
```

Out[418...

| | ProductName | AverageRating |
|---|--|---------------|
| 0 | Ariel Matic Top Load Liquid Detergent | 3.05 |
| 1 | Amul Taaza Toned Milk | 3.05 |
| 2 | Tata Tea Chakra Gold Premium Tea | 3.05 |
| 3 | Tata Tea | 3.04 |
| 4 | Nimyle Herbal Anti-Bacterial Floor Cleaner | 3.04 |

```
In [25]: # Identify the total sales by each category ?
query = """SELECT Category, round(SUM(Total_Order_Value)) AS TotalRevenue
FROM dmartready
GROUP BY Category order by TotalRevenue desc"""
cursor.execute(query)
data5b = cursor.fetchall()
df5b = pd.DataFrame(data5b, columns = ["Category", "TotalSales"])
df5b.head()
```

```
Out[25]:
```

| | Category | TotalSales |
|---|----------|------------|
| 0 | Local | 8482035.0 |
| 1 | Branded | 5124795.0 |
| 2 | Imported | 3347279.0 |

```
In [73]: # Total Sales by state ?
query = """SELECT State, round(SUM(Total_Order_Value)) AS TotalRevenue
FROM dmartready
GROUP BY State order by TotalRevenue desc"""
cursor.execute(query)
data5c = cursor.fetchall()
df5c = pd.DataFrame(data5c, columns = ["State", "TotalSales"])
df5c
```

```
Out[73]:
```

| | State | TotalSales |
|---|----------------|------------|
| 0 | Maharashtra | 4298237.0 |
| 1 | Andhra Pradesh | 4249018.0 |
| 2 | Gujarat | 4225412.0 |
| 3 | Telangana | 4181442.0 |

```
In [366... # Create a Donut chart using Plotly
fig = px.pie(df5c,
              names='State', # Use the 'State' column for the segments
              values='TotalSales', # Use 'TotalSales' as the size of each segment
              title="Total Sales by State",
              template="plotly_dark",
              hole=0.5) # This creates the donut shape by making a hole in the center

# Show the interactive plot
fig.show()
```

```
In [128... # Total Sales by top 10 City ?
query = """SELECT City, round(SUM(Total_Order_Value)) AS TotalRevenue
FROM dmartready
GROUP BY City order by TotalRevenue desc"""
cursor.execute(query)
data5d = cursor.fetchall()
df5d = pd.DataFrame(data5d, columns = ["City", "TotalSales"])
df5d.head(10)
```

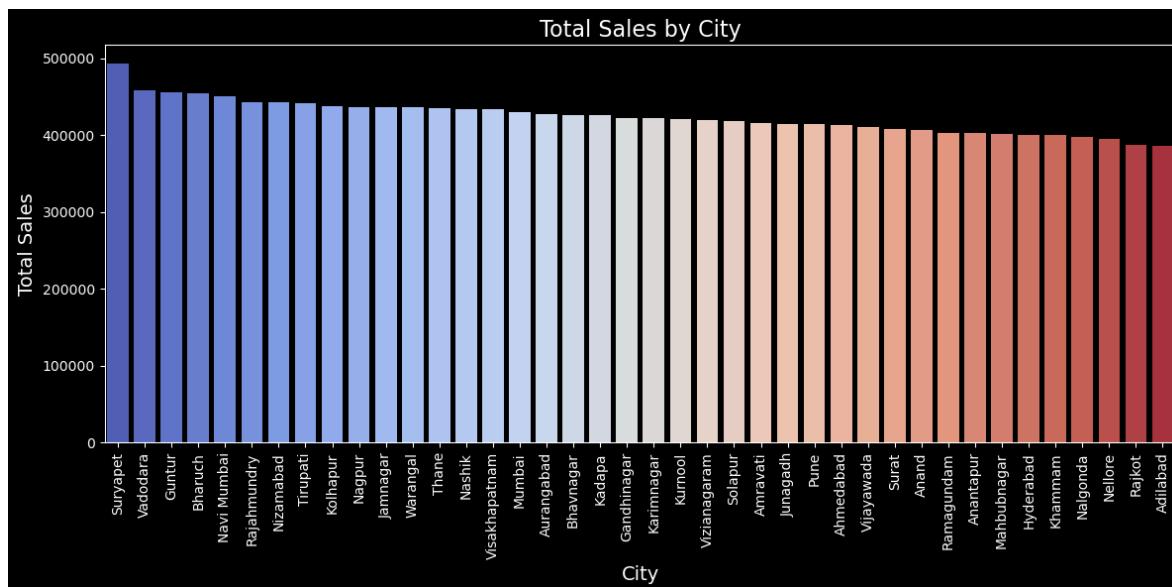
Out[128...

| | City | TotalSales |
|---|-------------|------------|
| 0 | Suryapet | 492450.0 |
| 1 | Vadodara | 457906.0 |
| 2 | Guntur | 456108.0 |
| 3 | Bharuch | 453977.0 |
| 4 | Navi Mumbai | 450947.0 |
| 5 | Rajahmundry | 443223.0 |
| 6 | Nizamabad | 442528.0 |
| 7 | Tirupati | 441999.0 |
| 8 | Kolhapur | 436959.0 |
| 9 | Nagpur | 436765.0 |

In [352...

```
# Set the figure size for the plot
plt.figure(figsize=(12, 6))

# Plotting the bar chart
sns.barplot(x='City', y='TotalSales', data=df5d, hue = 'City', palette='coolwarm')
plt.xticks(rotation=90)
plt.title('Total Sales by City', fontsize=16)
plt.xlabel('City', fontsize=14)
plt.ylabel('Total Sales', fontsize=14)
plt.tight_layout() # Adjust the plot to fit everything
plt.show()
```



In [308...

```
# Identify Operational Efficiency for delivery?
# Convert Columns to Datetime
df['OrderDate'] = pd.to_datetime(df['OrderDate'], format='%d-%m-%Y')
df['DeliveryDate'] = pd.to_datetime(df['DeliveryDate'], format='%d-%m-%Y')
# Calculate DeliveryTime
df['DeliveryTime'] = (df['DeliveryDate'] - df['OrderDate']).dt.days
# Check for DeliveryTime column
print(df[['OrderDate', 'DeliveryDate', 'DeliveryTime']])
```

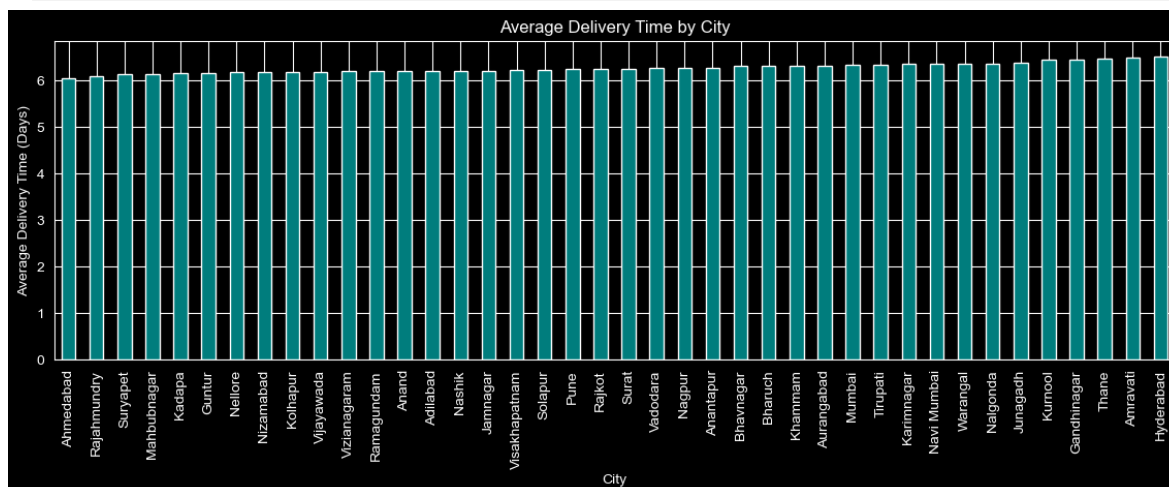
```
# Calculate average delivery time by city
avg_delivery_time = df.groupby('City')['DeliveryTime'].mean()
```

| | OrderDate | DeliveryDate | DeliveryTime |
|-------|------------|--------------|--------------|
| 0 | 2022-05-05 | 2022-05-13 | 8 |
| 1 | 2021-08-13 | 2021-08-20 | 7 |
| 2 | 2021-10-17 | 2021-10-26 | 9 |
| 3 | 2021-08-25 | 2021-08-27 | 2 |
| 4 | 2023-12-08 | 2023-12-10 | 2 |
| ... | ... | ... | ... |
| 24995 | 2021-04-02 | 2021-04-04 | 2 |
| 24996 | 2021-04-10 | 2021-04-13 | 3 |
| 24997 | 2023-09-01 | 2023-09-05 | 4 |
| 24998 | 2023-07-13 | 2023-07-23 | 10 |
| 24999 | 2022-10-26 | 2022-11-05 | 10 |

[25000 rows x 3 columns]

In [430...

```
# Plotting
plt.style.use('dark_background')
plt.figure(figsize=(14, 4))
avg_delivery_time.sort_values().plot(kind='bar', color='teal')
plt.title('Average Delivery Time by City')
plt.xlabel('City')
plt.ylabel('Average Delivery Time (Days)')
plt.xticks(rotation=90)
plt.show()
```



In [434...

```
# Total sales by each types of members ?
query = """SELECT Subscription, round(SUM(Total_Order_Value)) AS TotalRevenue
FROM dmartready
GROUP BY Subscription order by TotalRevenue desc"""
cursor.execute(query)
data5e = cursor.fetchall()
df5e = pd.DataFrame(data5e, columns = ["SubscriptionCategory", "TotalSales"])
df5e.head()
```

Out[434...

| | SubscriptionCategory | TotalSales |
|---|----------------------|------------|
| 0 | Freepass | 9720905.0 |
| 1 | Premium | 5173185.0 |
| 2 | Premium Plus | 2060020.0 |


```
In [436... fig = px.pie(df5e,
               names='SubscriptionCategory', values='TotalSales', title="Total Sal
               color='SubscriptionCategory', # Color the slices by the MarketingP
               hover_data=['TotalSales'], # Show the total sales on hover
               labels={'TotalSales': 'TotalSales', 'SubscriptionCategory': 'Subscr
               template="plotly_dark" # theme to 'plotly_dark' for a dark backgro
               )
fig.show()
```

```
In [67]: # Total sales by each marketing campaign ?
query = """SELECT `Marketing/Advertisement`, round(SUM(Total_Order_Value)) AS To
FROM dmartready
GROUP BY `Marketing/Advertisement` order by TotalRevenue desc"""
cursor.execute(query)
data5f = cursor.fetchall()
df5f = pd.DataFrame(data5f, columns = ["MarketingPlatform", "TotalSales"])
df5f.head()
```

Out[67]:

| | MarketingPlatform | TotalSales |
|---|-------------------|------------|
| 0 | Instagram | 5755979.0 |
| 1 | Facebook | 4219599.0 |
| 2 | Other | 2803532.0 |
| 3 | Friends | 2779127.0 |
| 4 | TV | 1395873.0 |

In [326...

```
fig = px.pie(df5f,
              names='MarketingPlatform', values='TotalSales', title="Total Sales",
              color='MarketingPlatform', # Color the slices by the MarketingPlatform
              hover_data=['TotalSales'], # Show the total sales on hover
              labels={'TotalSales': 'Total Sales', 'MarketingPlatform': 'MarketingPlatform'},
              template="plotly_dark" # theme to 'plotly_dark' for a dark background
            )
fig.show()
```

In [202...

```
# Find the average time spent on the website by customers who made a purchase in
query = """select state, round(avg(time_spent_on_website),2) as AverageTimeSpent
cursor.execute(query)
data6 = cursor.fetchall()
df6 = pd.DataFrame(data6, columns = ["State", "ProductPurchased"])
df6.head()
```

Out[202...

| | State | AverageTimeSpent |
|---|----------------|------------------|
| 0 | Gujarat | 9.99 |
| 1 | Maharashtra | 10.10 |
| 2 | Telangana | 10.26 |
| 3 | Andhra Pradesh | 10.50 |

In [446...

```
# Find the average time spent on the website by customers who made a purchase in
query = """select city, round(avg(time_spent_on_website),2) as AverageTimeSpent
order by AverageTimeSpent desc"""
cursor.execute(query)
data7 = cursor.fetchall()
df7 = pd.DataFrame(data7, columns = ["City", "AverageTimeSpent"])
df7.head()
```

Out[446...

| | City | AverageTimeSpent |
|---|------------|------------------|
| 0 | Ramagundam | 11.69 |
| 1 | Vijayawada | 11.50 |
| 2 | Karimnagar | 11.34 |
| 3 | Junagadh | 11.18 |
| 4 | Rajkot | 11.17 |

In [452...

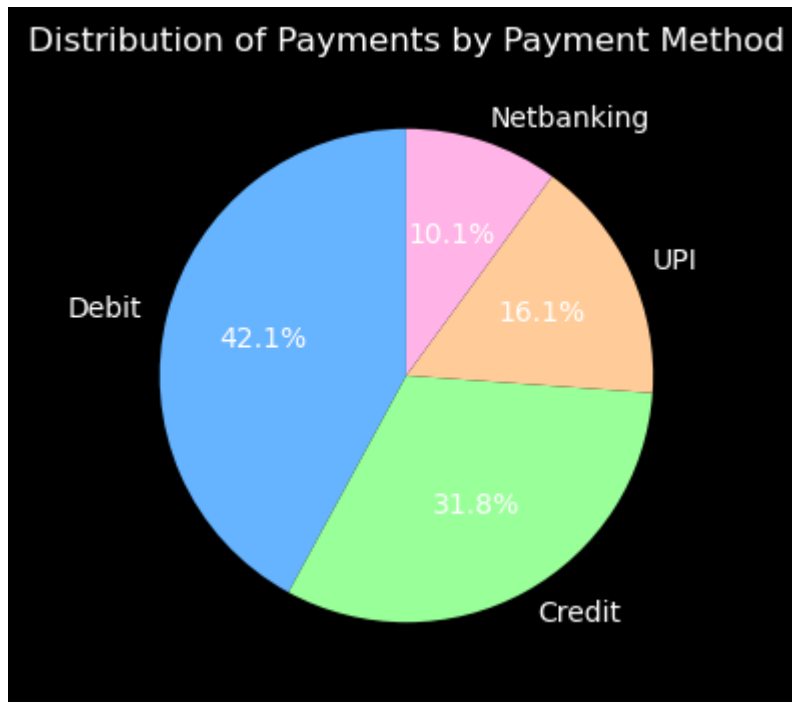
```
# Find the total sales (OrderValue) by payment method.
query = """SELECT Payment_Method, round(SUM(Total_Order_Value)) AS TotalSales
FROM dmartready
GROUP BY Payment_Method order by TotalSales desc;"""
cursor.execute(query)
data8 = cursor.fetchall()
df8 = pd.DataFrame(data8, columns = ["PaymentMethod", "TotalPayment"])
df8a = df8.head()
df8a
```

Out[452...

| | PaymentMethod | TotalPayment |
|---|---------------|--------------|
| 0 | Debit | 6778049.0 |
| 1 | Credit | 5124057.0 |
| 2 | UPI | 2587674.0 |
| 3 | Netbanking | 1620846.0 |
| 4 | COD | 843485.0 |

In [318...

```
# Create a pie chart
plt.figure(figsize=(4, 4))
plt.pie(df8a['TotalPayment'], labels=df8a['PaymentMethod'], autopct='%1.1f%%', s
plt.title('Distribution of Payments by Payment Method')
plt.show()
```



```
In [272... # Find the products with the highest average time spent on the website by custom
query = """SELECT Product_Name, round(AVG(Time_Spent_on_Website),2) AS AverageTi
FROM dmartready GROUP BY Product_Name
ORDER BY AverageTimeSpent DESC"""
cursor.execute(query)
data9 = cursor.fetchall()
df9 = pd.DataFrame(data9, columns = ["ProductName", "AverageTimeSpent"])
df9.head()
```

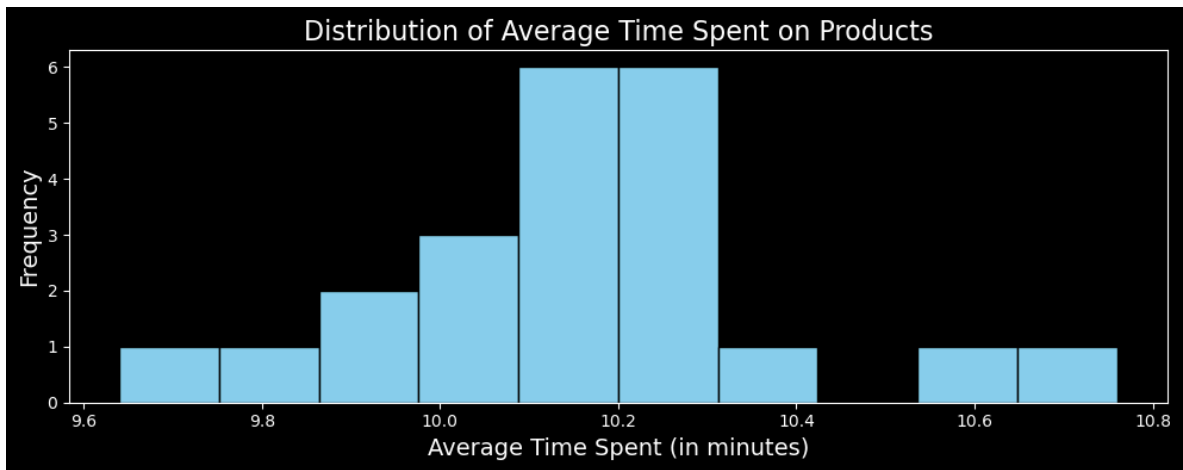
```
Out[272...      ProductName  AverageTimeSpent
0  Ariel Matic Top Load Liquid Detergent    10.76
1      Amul Taaza Toned Milk    10.58
2  Cavin's Chocolate Milkshake    10.36
3      Tata Sampann Chana Dal    10.30
4  Tata Sampann Moong Dal (Pesara Pappu)    10.26
```

```
In [340... # Apply dark background style
plt.style.use('dark_background')

# Create a histogram with a dark background
plt.figure(figsize=(10, 4))
plt.hist(df9['AverageTimeSpent'], bins=10, color='skyblue', edgecolor='black')

# Add titles and Labels
plt.title('Distribution of Average Time Spent on Products', fontsize=16)
plt.xlabel('Average Time Spent (in minutes)', fontsize=14)
plt.ylabel('Frequency', fontsize=14)

# Show the plot
plt.tight_layout()
plt.show()
```



```
In [462... # Which states & city have the highest cancellation rates (orders cancelled / to
query = """SELECT State, city,
            (COUNT(Cancellation_Date) / COUNT(Order_ID)) * 100 AS CancellationRate
FROM dmartready GROUP BY State, city
ORDER BY state, CancellationRate DESC;"""
cursor.execute(query)
data10 = cursor.fetchall()
df10 = pd.DataFrame(data10, columns = ["State", "City", "CancellationRate"])
df10.head()
```

```
Out[462...
      State      City  CancellationRate
0  Andhra Pradesh  Anantapur          8.7302
1  Andhra Pradesh   Tirupati          8.1340
2  Andhra Pradesh  Vijayawada          8.0725
3  Andhra Pradesh   Guntur           7.9646
4  Andhra Pradesh  Rajahmundry          7.5617
```

```
In [490... data = {
    'State': ['Andhra Pradesh'] * 10 + ['Gujarat'] * 10 + ['Maharashtra'] * 10 +
    'City': ['Anantapur', 'Tirupati', 'Vijayawada', 'Guntur', 'Rajahmundry', 'Vi
            'Bhavnagar', 'Ahmedabad', 'Rajkot', 'Vadodara', 'Surat', 'Anand', '
            'Kolhapur', 'Thane', 'Amravati', 'Aurangabad', 'Solapur', 'Pune', '
            'Ramagundam', 'Nalgonda', 'Hyderabad', 'Adilabad', 'Nizamabad', 'Kh
    'CancellationRate': [8.7302, 8.1340, 8.0725, 7.9646, 7.5617, 7.1537, 6.8376,
                        7.7922, 7.6800, 7.6792, 7.6583, 7.5353, 6.9307, 6.8910,
                        10.2524, 10.2326, 8.0569, 7.6205, 7.5929, 7.2968, 6.637
                        7.9077, 7.1186, 6.9652, 6.7241, 6.5625, 6.4570, 6.4364,
    }

dfdata = pd.DataFrame(data)

# Create the interactive bar plot
fig = px.bar(dfdata,
             x='City',
             y='CancellationRate',
             color='State',
             title='Cancellation Rate by City and State',
             labels={'CancellationRate': 'Cancellation Rate (%)'},
             template="plotly_dark",
             hover_data=['State'])
```

```
fig.update_layout(barmode='group', xaxis_tickangle=-45)
fig.show()
```

```
In [472... # Total sales in each year ?
query = """select year, sum(Total_Order_Value) as TotalSales from dmartready gro
cursor.execute(query)
data11c = cursor.fetchall()
df11c = pd.DataFrame(data11c, columns = ["Year", "TotalSales"])
df11c.head()
```

```
Out[472...
   Year  TotalSales
0  2021  5526504.64
1  2022  5729150.56
2  2023  5698454.13
```

```
In [496... # Creating a box plot with Plotly Express
fig = px.box(df, x="Year", y="TotalOrderValue", title="Total Sales Distribution
            labels={"TotalSales": "Total Sales", "Year": "Year"},
            template="plotly_dark")

# Show the plot
fig.show()
```

```

In [512... # Total sales by each year & month ?
query = """select year, month, round(sum(Total_Order_Value)) as TotalSales from
cursor.execute(query)
data11d = cursor.fetchall()
df11d = pd.DataFrame(data11d, columns = ["Year", "Month", "TotalSales"])
# Define the correct month order
month_order = ['January', 'February', 'March', 'April', 'May', 'June', 'July', '
# Convert the 'Month' column to a categorical type with the specified order
df11d['Month'] = pd.Categorical(df11d['Month'], categories=month_order, ordered=

# Sort the DataFrame based on 'Year' and 'Month'
df_sorted = df11d.sort_values(by=['Year', 'Month'])

# Display the sorted DataFrame
df_sorted.head()

```

```

Out[512...

```

| | Year | Month | TotalSales |
|---|------|----------|------------|
| 4 | 2021 | January | 450819.0 |
| 3 | 2021 | February | 408273.0 |
| 7 | 2021 | March | 462025.0 |
| 0 | 2021 | April | 466984.0 |
| 8 | 2021 | May | 461050.0 |

```

In [76]: # Plotting the data using Plotly Express
fig = px.line(df_sorted, x="Month", y="TotalSales", color="Year", title="Total S
          labels={"TotalSales": "Total Sales", "Month": "Month", "Year": "Ye
          markers=True)

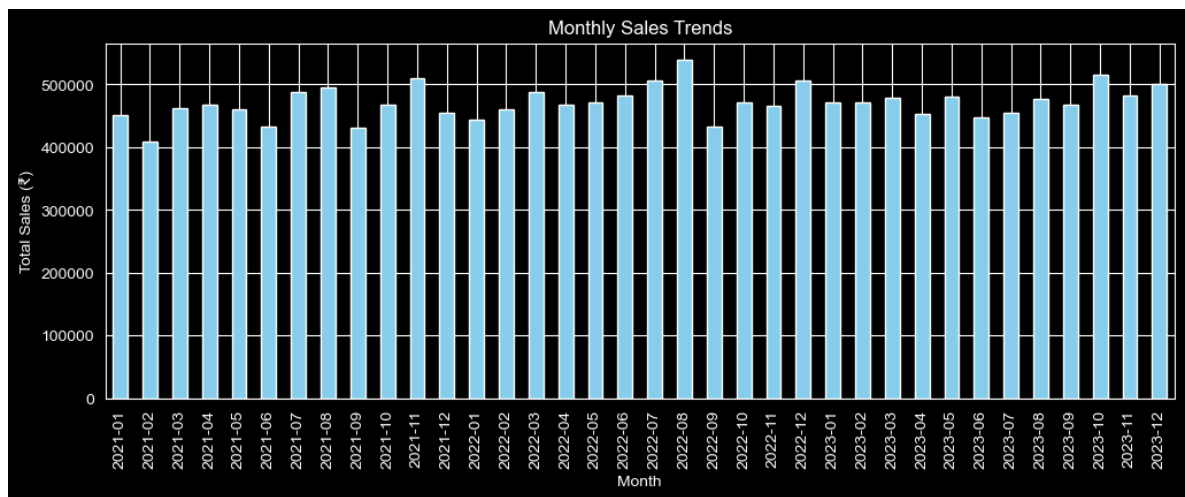
# Update Layout for dark background
fig.update_layout(
    plot_bgcolor='rgb(32, 32, 32)', # Dark background for the plot area
    paper_bgcolor='rgb(28, 28, 28)', # Dark background for the entire figure
    font=dict(color='white'), # Set text color to white for better contrast
    title_font=dict(size=20, color='white'), # Title font color
    xaxis=dict(showgrid=True, gridcolor='rgb(50, 50, 50)', title_font=dict(color
    yaxis=dict(showgrid=True, gridcolor='rgb(50, 50, 50)', title_font=dict(color
)
# Show the plot
fig.show()

```

```

In [526... # Sales Trends
plt.style.use('dark_background')
sales_trends = df.groupby(df['OrderDate'].dt.to_period("M"))['TotalOrderValue'].
plt.figure(figsize=(12, 4))
sales_trends.plot(kind='bar', color='skyblue', title='Monthly Sales Trends')
plt.xlabel('Month')
plt.ylabel('Total Sales (₹)')
plt.show()

```

```
In [222... # Analyze the delivery delay by calculating the difference between the order dat
query = """SELECT Order_ID, DATEDIFF(Delivery_Date, Order_Date) AS DeliveryDelay
FROM dmartready
WHERE Delivery_Date IS NOT NULL"""
cursor.execute(query)
data12 = cursor.fetchall()
df12 = pd.DataFrame(data12, columns = ["OrderID", "DeliveryDelay"])
df12
```

```
Out[222...      OrderID  DeliveryDelay
0  479577309             None
1  634865221             None
2  113166210             None
3  740539230             None
4  156544145             None
...      ...             ...
24995  449449740             None
24996  171612408             None
24997  961751448             None
24998  258727912             None
24999  962056465             None
```

25000 rows × 2 columns

```
In [528... # Identify the top 10 products with the highest discount rate (based on the perc
query = """SELECT product_name, (Discount_Price / MRP) * 100 AS DiscountPercenta
FROM dmartready
ORDER BY DiscountPercentage DESC
LIMIT 10"""
cursor.execute(query)
data13 = cursor.fetchall()
df13 = pd.DataFrame(data13, columns = ["ProductName", "DiscountPercentage"])
df13r = df13.round()
df13r
```

Out[528...

| | ProductName | DiscountPercentage |
|---|--|--------------------|
| 0 | Nimyle Herbal Anti-Bacterial Floor Cleaner | 95.0 |
| 1 | Nimyle Herbal Anti-Bacterial Floor Cleaner | 95.0 |
| 2 | Usha Heavy Weight Iron - EI3710: 1000 W | 95.0 |
| 3 | Premia Tur Dal Latur (Kandi Pappu) | 95.0 |
| 4 | Tata Tea Chakra Gold Premium Tea | 95.0 |
| 5 | Wagh Bakri Masala Tea | 95.0 |
| 6 | Nutri Delite Breakfast Cereal | 95.0 |
| 7 | Tata Tea Chakra Gold Premium Tea | 95.0 |
| 8 | Tata Tea Chakra Gold Premium Tea | 95.0 |
| 9 | Pigeon Induction Cooktop Acer Plus: 1800 W | 95.0 |

In [468...

```

# Group by ProductName and calculate the mean DiscountPercentage
df_grouped = df13.groupby('ProductName', as_index=False)['DiscountPercentage'].mean()

# Create the bar chart
fig = px.bar(df_grouped,
             x='ProductName',
             y='DiscountPercentage',
             title='Discount Percentage by Product',
             labels={'DiscountPercentage': 'Discount Percentage', 'ProductName': 'Product Name'},
             color='DiscountPercentage', # Optional: Add color to make the chart more readable
             template="plotly_dark",
             text='DiscountPercentage' # Show DiscountPercentage on top of the bars
            )

# Rotate x-axis labels for better readability
fig.update_xaxes(tickangle=25)

# Show the plot
fig.show()

```

```
In [234... # Identify the sales based on gender ?
query = """select gender, round(sum(total_order_value)) as TotalSales from dmart
cursor.execute(query)
data14 = cursor.fetchall()
df14 = pd.DataFrame(data14, columns = ["Gender", "Total"])
df14
```

```
Out[234...

```

| | Gender | Total |
|---|--------|-----------|
| 0 | Male | 9688741.0 |
| 1 | Female | 7265369.0 |

```
In [240... # Identify the product wise shipping charges ?
query = """select product_name, sum(shipping_charges) as TotalShippingCharges fr
cursor.execute(query)
data15 = cursor.fetchall()
df15 = pd.DataFrame(data15, columns = ["ProductName", "TotalShippingCharges"])
df15.head()
```

Out[240...

| | ProductName | TotalShippingCharges |
|---|---------------------------------------|----------------------|
| 0 | Ariel Matic Top Load Liquid Detergent | 42325.0 |
| 1 | VSR Channa Dal | 38875.0 |
| 2 | Tenali Double Horse Chana Dal | 43100.0 |
| 3 | Tata Tea | 43025.0 |
| 4 | Sofit Soya Drink Vanilla | 38475.0 |

In [538...

```
# Identify the shipmode wise total sales based on order status ?
query = """select ship_mode, order_status, round(sum(total_order_value),2) as To
order_status order by TotalRevenue desc"""
cursor.execute(query)
data16 = cursor.fetchall()
df16 = pd.DataFrame(data16, columns = ["ShipMode", "OrderStatus", "TotalRevenue"]
df16.head(6)
```

Out[538...

| | ShipMode | OrderStatus | TotalRevenue |
|---|--------------|-------------|--------------|
| 0 | Free | Delivered | 5385100.22 |
| 1 | Free | Shipped | 3753593.82 |
| 2 | Priority | Delivered | 2580485.54 |
| 3 | Express plus | Delivered | 2083202.70 |
| 4 | Priority | Shipped | 1794704.86 |
| 5 | Express plus | Shipped | 1357022.19 |

In [546...

```
# Marketing Effectiveness
# Identify the total revenue based on (gender & marketing/advertisement) ?
query = """select gender, `Marketing/Advertisement`, sum(Total_Order_Value) as T
group by gender, `Marketing/Advertisement` order by TotalRevenue desc"""
cursor.execute(query)
data17 = cursor.fetchall()
df17 = pd.DataFrame(data17, columns = ["Gender", "Marketing/Advertisement", "Tot
df17.head(10)
```

Out[546...

| | Gender | Marketing/Advertisement | TotalRevenue |
|---|--------|-------------------------|--------------|
| 0 | Male | Instagram | 3280998.42 |
| 1 | Female | Instagram | 2474980.60 |
| 2 | Male | Facebook | 2430219.62 |
| 3 | Female | Facebook | 1789379.15 |
| 4 | Male | Other | 1582572.71 |
| 5 | Male | Friends | 1565633.57 |
| 6 | Female | Other | 1220959.05 |
| 7 | Female | Friends | 1213493.08 |
| 8 | Male | TV | 829316.24 |
| 9 | Female | TV | 566556.89 |

In [544...

```
# Create an interactive Sunburst chart
fig = px.sunburst(df17,
                  path=['Gender', 'Marketing/Advertisement'], # Hierarchical levels
                  values='TotalRevenue', # Values that determine the size of the segments
                  title="Total Revenue by Gender and Marketing Platform", # Title
                  template="plotly_dark",
                  labels={"TotalRevenue": "Total Revenue", "Gender": "Gender", "Marketing/Advertisement": "Marketing Platform"})

# Show the interactive plot
fig.show()
```

```

In [342... # Profit by each category ?
# Replace "Cancelled" and "Returned" in the 'Order_Status' column with 0
df['TotalOrderValue'] = df['TotalOrderValue'].where(~df['TotalOrderValue'].isin(

In [346... df['ShippingCharges'] = df['ShippingCharges'].where(~df['ShippingCharges'].isin(

In [352... df['TotalOrderValue'] = pd.to_numeric(df['TotalOrderValue'], errors='coerce')
df['MRP'] = pd.to_numeric(df['MRP'], errors='coerce')
df['DiscountPrice'] = pd.to_numeric(df['DiscountPrice'], errors='coerce')
df['ShippingCharges'] = pd.to_numeric(df['ShippingCharges'], errors='coerce')

In [354... df['Profit'] = df['TotalOrderValue'] - (df['MRP'] - df['DiscountPrice'] + df['Sh

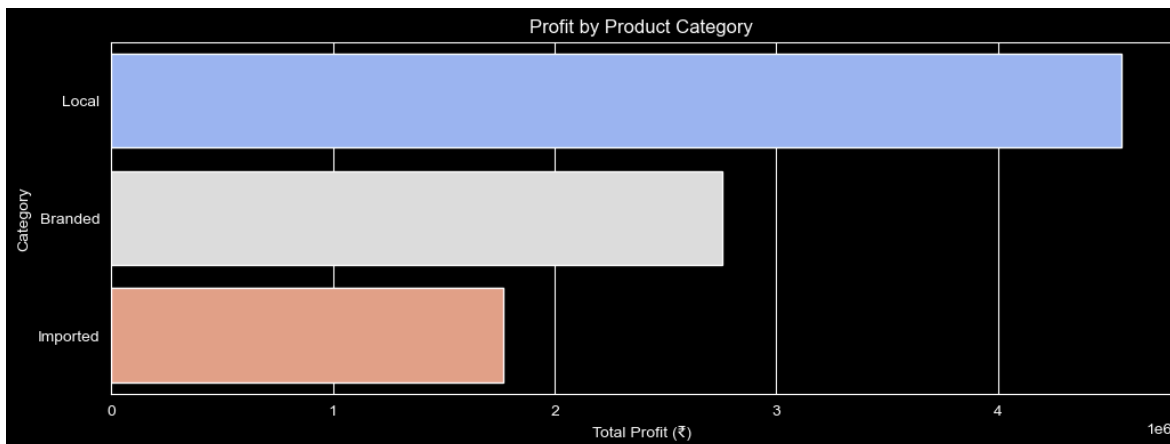
In [390... profit_by_category = df.groupby('Category')['Profit'].sum().sort_values(ascending
profit_by_category

Out[390... Category
Local      4557246.29
Branded    2754302.54
Imported   1769457.06
Name: Profit, dtype: float64

In [392... # Profitability Analysis
plt.style.use('dark_background')
plt.figure(figsize=(12, 4))
sns.barplot(x=profit_by_category.values, y=profit_by_category.index, palette='co
plt.title('Profit by Product Category')

```

```
plt.xlabel('Total Profit (₹)')
plt.ylabel('Category')
plt.show()
```



In [388...

```
summary = {
    'Metric': ['Highest Sales Month', 'Top-Selling Product', 'Best Customer Segment', 'Fastest City'],
    'Value': [
        sales_trends.idxmax(),
        product_popularity.idxmax(),
        customer_segmentation.stack().idxmax(),
        avg_delivery_time.idxmin()
    ]
}
summary_df = pd.DataFrame(summary)
print("Summary of Key Insights")
print(summary_df)
```

Summary of Key Insights

| | Metric | Value |
|---|-----------------------|----------------------------------|
| 0 | Highest Sales Month | 2022-08 |
| 1 | Top-Selling Product | Harpic Bathroom Cleaner - Floral |
| 2 | Best Customer Segment | (Male, 36-45) |
| 3 | Fastest City | Ahmedabad |

Reporting

Sales Insights

Total Sales: The store generated total sales of ₹16,954,109 over the analyzed period. Average sales per order are ₹678.16, indicating moderate order values.

Yearly Performance: Sales have remained steady across years, with slight peaks in 2022 at ₹5,729,150.56. Orders have grown from 8,310 in 2021 to 8,365 in 2023, suggesting slight growth in demand.

Monthly Trends: November and December consistently show higher sales and orders, likely due to festive shopping. February and September have lower sales and order volumes across years, indicating potential off-seasons.

Category Performance: Local products contribute the most to sales (₹8,482,035) and profit (₹4,557,246.29), emphasizing customer preference for local goods. Branded products are the second-best performer.

Customer Behavior

Gender-Based Insights: Male customers contribute more to sales, especially in the 26-45 age group. Males are the primary audience for Instagram and Facebook campaigns.

State and City Analysis: Maharashtra, Andhra Pradesh, Gujarat, and Telangana are top-performing states, with Maharashtra leading sales at ₹4,298,237. Top cities like Suryapet, Vadodara, and Guntur have significant contributions to overall sales.

Time Spent on the Website: Andhra Pradesh customers spend the most time online (10.50 minutes on average). Cities like Ramagundam, Vijayawada, and Karimnagar have the highest engagement levels.

Product Performance

Top-Selling Products: Household essentials like Harpic Bathroom Cleaner and Parle Nutricrunch Digestive Cookies are top contributors to sales. Products with higher engagement also show high average time spent, such as Ariel Matic Top Load Liquid Detergent.

High Discount Products: Products like Nimyle Herbal Anti-Bacterial Floor Cleaner and Tata Tea Chakra Gold Premium Tea offer the highest discounts (95%), driving sales but potentially affecting margins.

Profitability: Local and branded products contribute significantly to profits, emphasizing a balance between cost and demand.

Operational Insights

Delivery Efficiency: Most deliveries occur within 2-10 days, with a need to improve delivery times for orders taking over a week. Free shipping generates the most revenue, but priority and express plus shipping are growing contributors.

Payment Preferences: Debit cards are the most popular payment method, followed by credit cards. COD (Cash on Delivery) contributes the least, indicating digital payment adoption.

Cancellation Rates: Andhra Pradesh and Gujarat cities like Anantapur and Bhavnagar have high cancellation rates, signaling potential issues with customer satisfaction or delivery logistics.

Marketing Effectiveness

Top Platforms: Instagram campaigns yield the highest revenue (₹5,755,979), especially among male customers. Facebook and referrals from friends also perform well, while TV ads are the least effective.

Gender-Specific Campaigns: Male customers engage more with Instagram, while female customers respond better to Facebook and referral campaigns.

Recommendations

Seasonal Campaigns: Focus on November and December with targeted marketing campaigns and attractive discounts to maximize festive demand.

Improve Delivery Times: Reduce delays for longer delivery periods to improve operational efficiency and customer satisfaction.

Optimize Discounts: Reassess high-discount strategies for products contributing low profitability while promoting profitable items.

City-Specific Strategies: Address high cancellation rates in cities like Anantapur and Bhavnagar by improving customer communication and delivery reliability.

Leverage Top Platforms: Double down on Instagram and Facebook campaigns for male audiences while exploring other platforms for female engagement.

Conclusion

The DmartReady Online Store Analysis highlights significant patterns and insights crucial for strategic decision-making. By leveraging data analysis tools like SQL and Python, the study provides actionable recommendations to enhance sales, improve customer satisfaction, and optimize operational efficiency. The findings underscore the importance of focusing on customer engagement, product diversification, and effective marketing strategies to maintain competitive advantage and drive long-term profitability.

Ultimately, this analysis serves as a roadmap for DmartReady to capitalize on its strengths and address areas of improvement, ensuring sustained growth in the competitive e-commerce landscape.

Thank You!

Questions or Feedback? Contact: @shreeram0912

In []: